

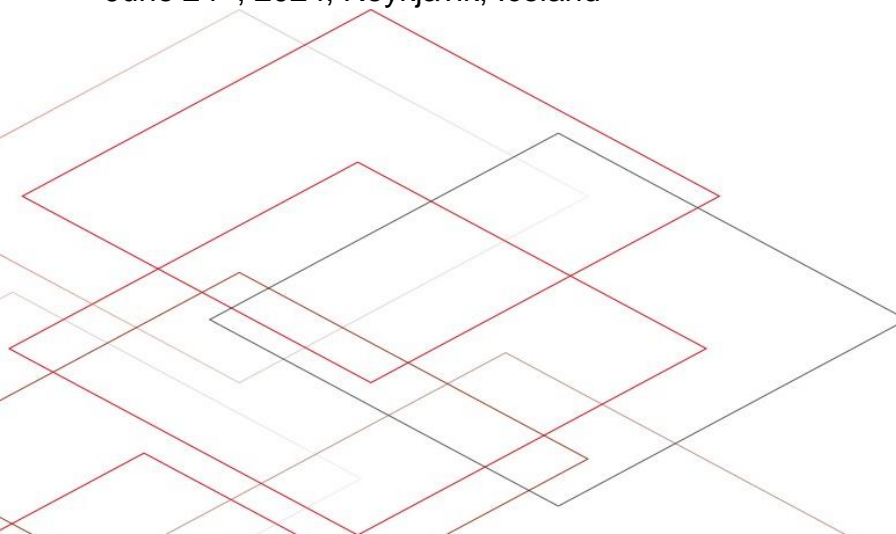
Introduction: Open Research Knowledge Graph

Oliver Karras, Alessio Ferrari, Davide Fucci, and Davide Dell'Anna

oliver.karras@tib.eu, alessio.ferrari@isti.cnr.it, davide.fucci@bth.se, d.dellanna@uu.nl

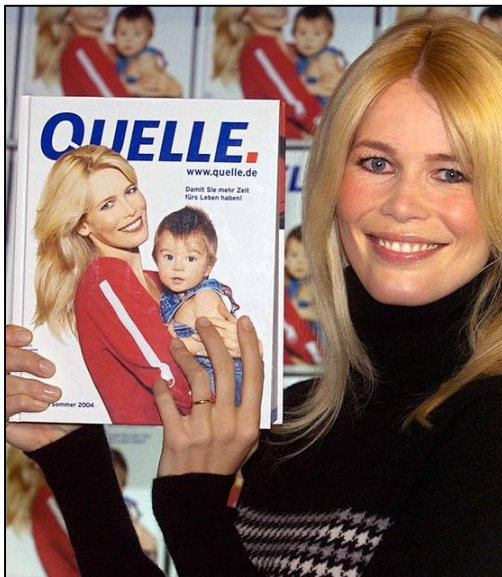
32nd IEEE International Requirements Engineering 2024 Conference – Exploring New Horizons: Expanding the Frontiers of Requirements Engineering

June 24th, 2024, Reykjavik, Iceland

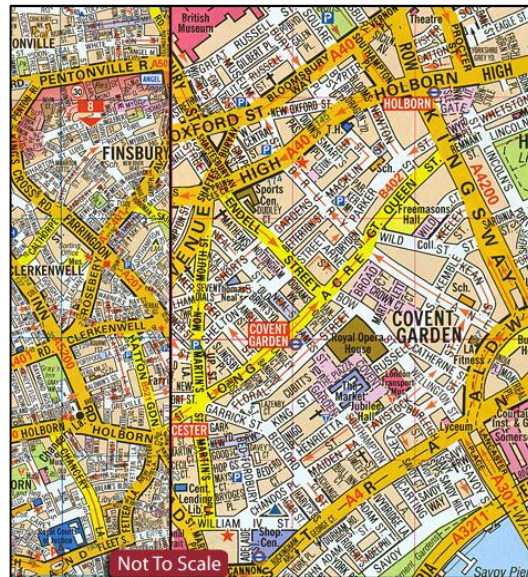


Once Upon a Time, we Communicated with Paper...

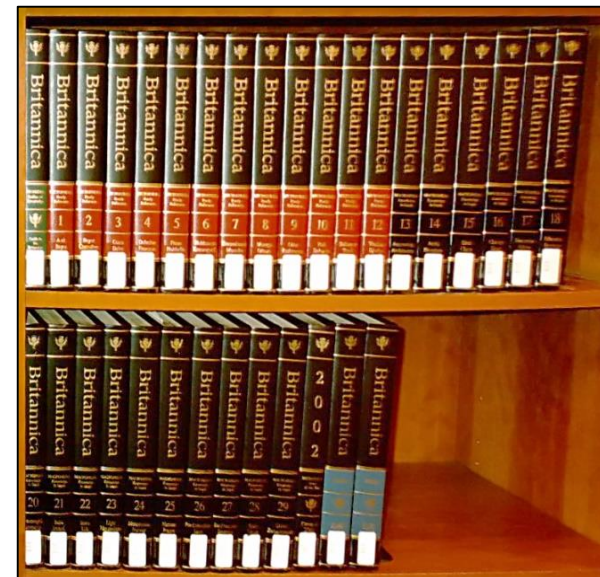
Who still remembers?



Mail order catalogs



Maps

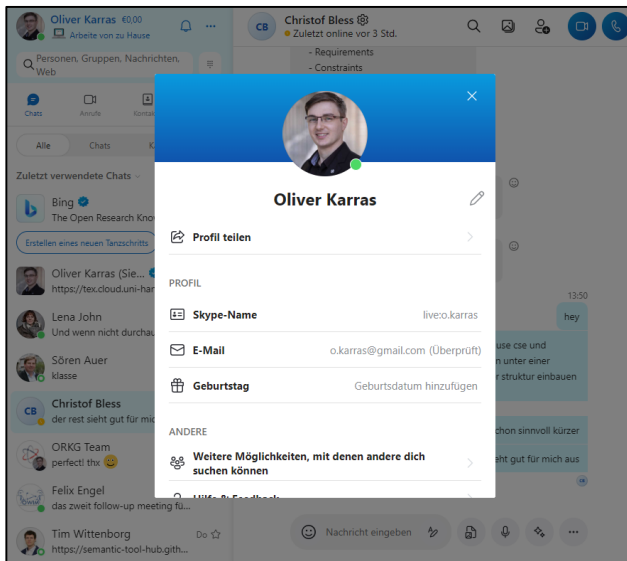
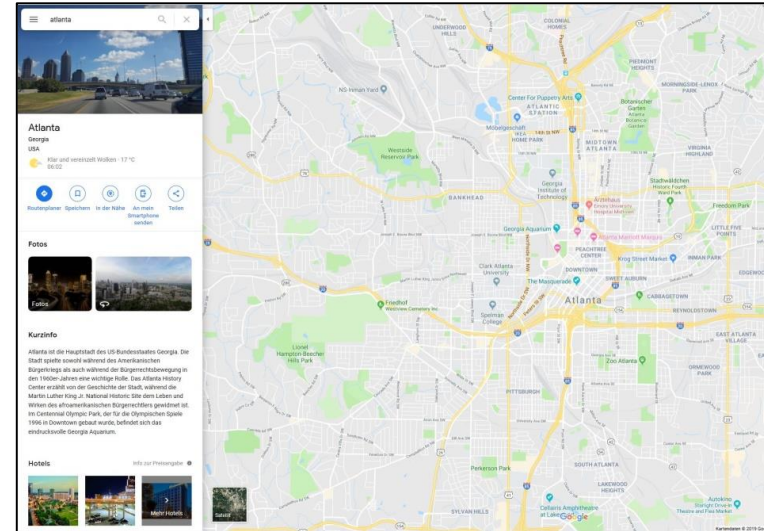
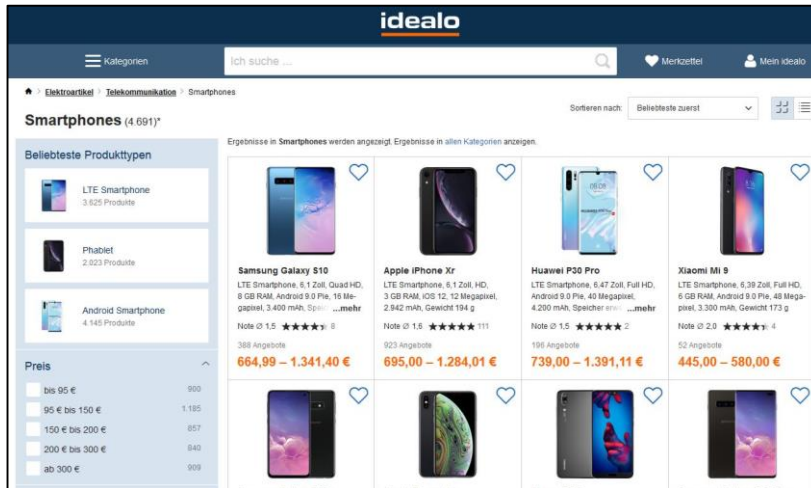


Encyclopedia



Phone books

... until Digital Transformation (Digitalization)!



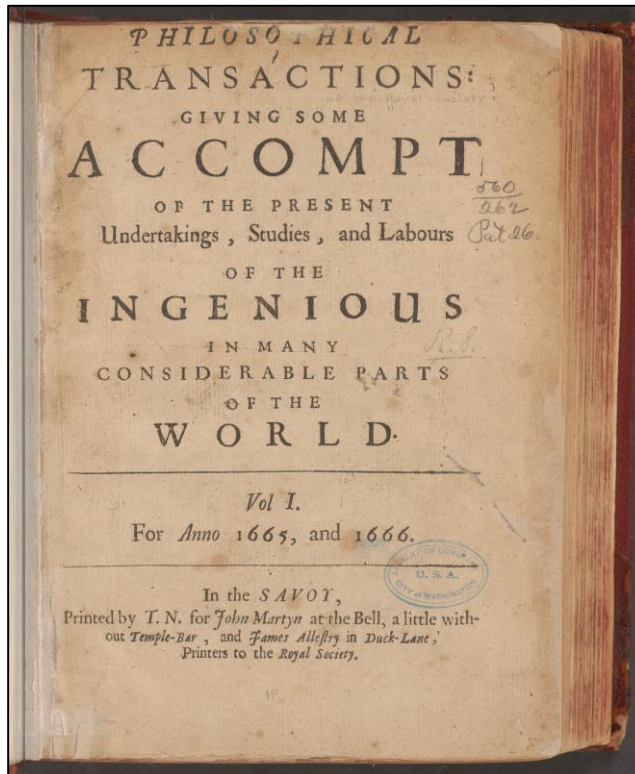
The World of Publishing & Communication has Profoundly Changed!

- **New means adapted to the new possibilities**, e.g., platforms
- **Completely new business models**
- **More focus on data, interlinking, services, and search**
- **Integration, crowdsourcing, and data curation** are important

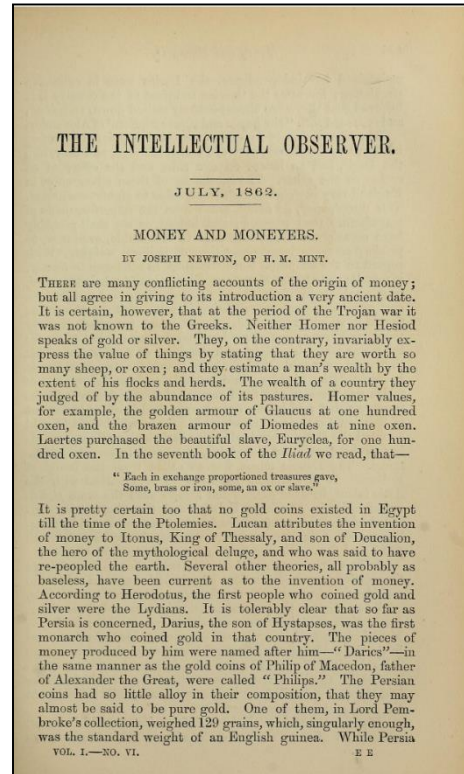
What has happened in academia in terms of scholarly publishing & communication?

Let's Take a Look

17th century



19th century



20th century

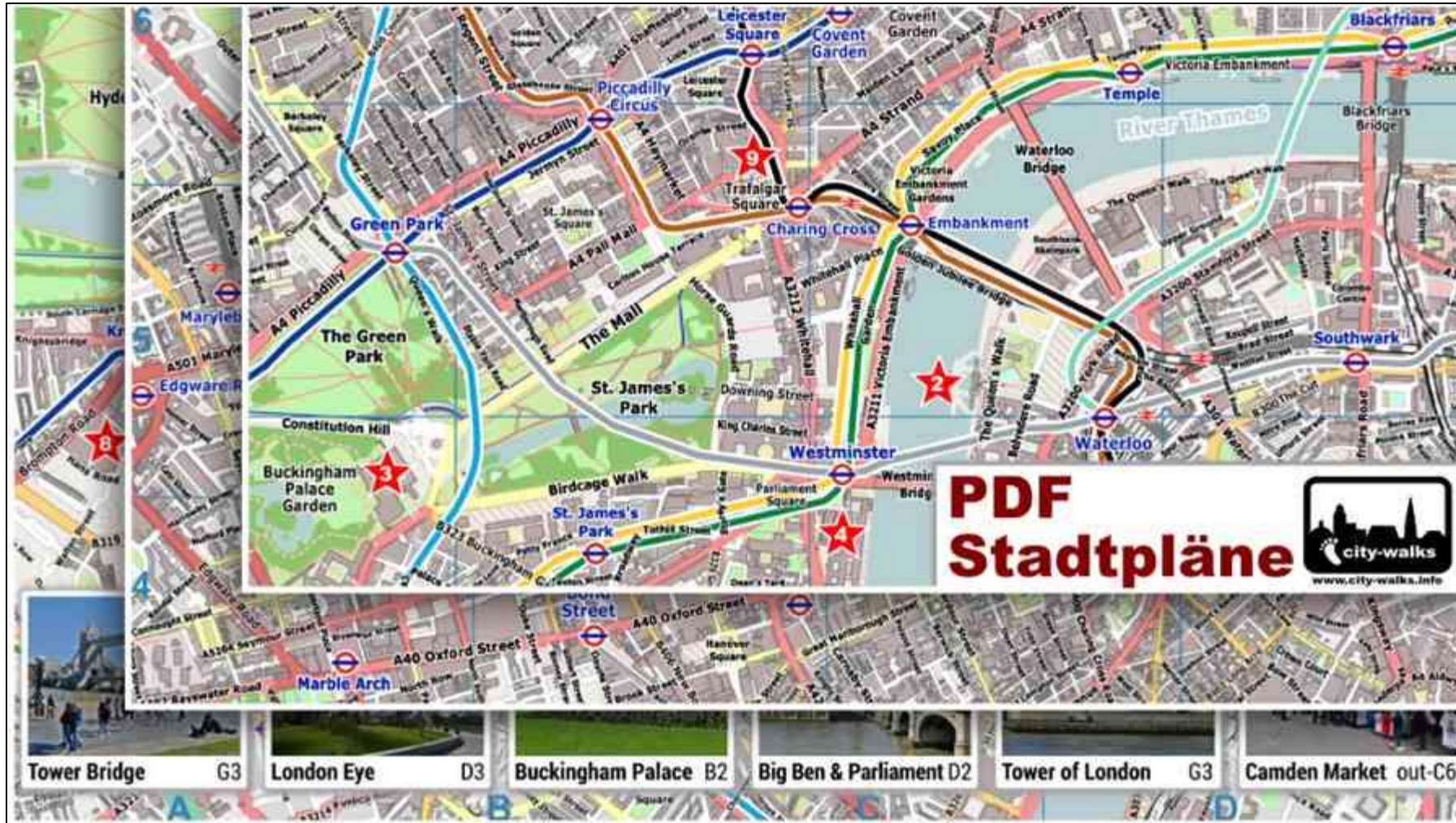


21st century



Scholarly publishing & communication has **not changed** (much)!

Let's Take a Look



21st century

AGDISTIS - Graph-Based Disambiguation of Named Entities Using Linked Data

Ricardo Usbeck^{1,2}, Axel-Cyrille Ngonga Ngomo¹, Michael Röder^{1,2},
Daniel Gerber¹, Sandro Athaide Coelho³, Sören Auer⁴, and Andreas Both²

¹ University of Leipzig, Germany

² R&D, Unister GmbH, Germany

³ Federal University of Juiz de Fora, Brazil

⁴ University of Bonn & Fraunhofer IAS, Germany
(usbeck,ngonga}@informatik.uni-leipzig.de

Abstract. Over the last decades, several billion Web pages have been made available on the Web. The ongoing transition from the current Web of unstructured data to the Web of Data yet requires scalable and accurate approaches for the extraction of structured data in RDF (Resource Description Framework) from these websites. One of the key steps towards extracting RDF from text is the disambiguation of named entities. While several approaches aim to tackle this problem, they still achieve poor accuracy. We address this drawback by presenting AGDISTIS, a novel knowledge-base-agnostic approach for named entity disambiguation. Our approach combines the Hypertext-Induced Topic Search (HITS) algorithm with label expansion strategies and string similarity measures. Based on this combination, AGDISTIS can efficiently detect the correct URIs for a given set of named entities within an input text. We evaluate our approach on eight different datasets against state-of-the-art named entity disambiguation frameworks. Our results indicate that we outperform the state-of-the-art approach by up to 29% F-measure.

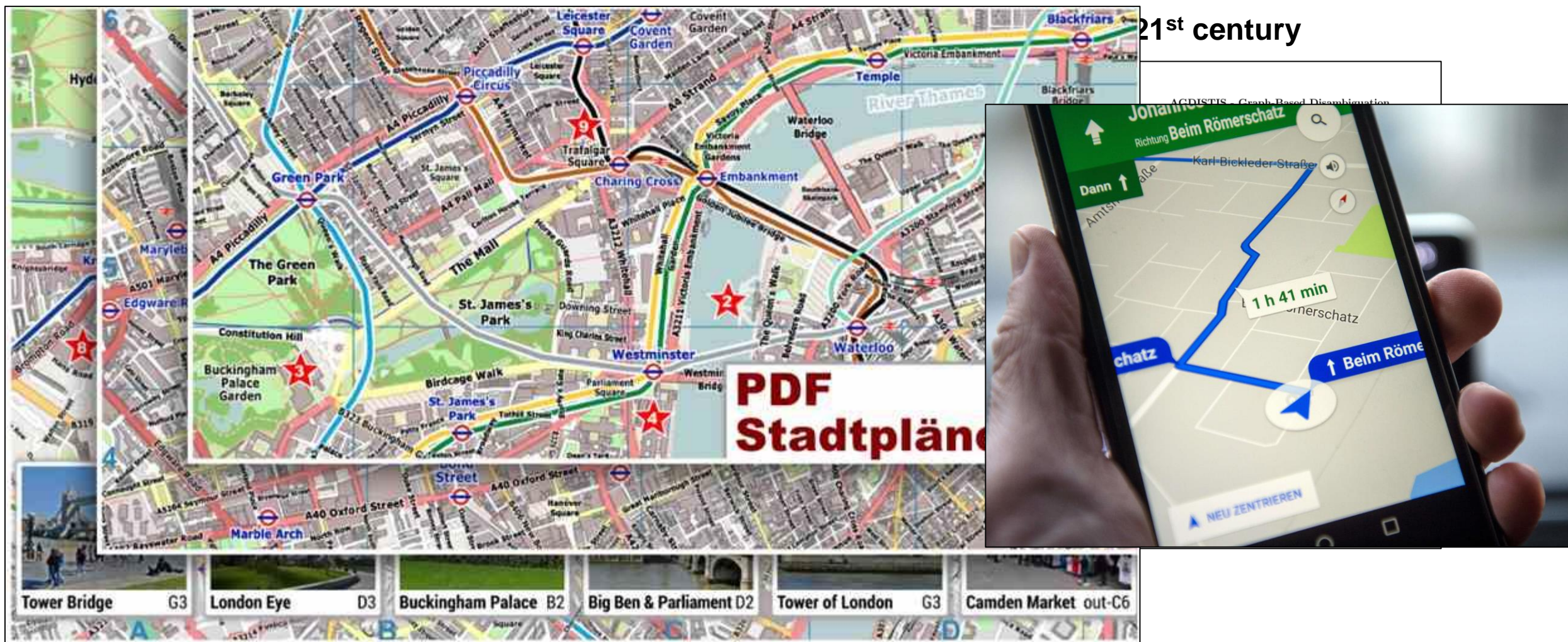
1 Introduction

The vision behind the Web of Data is to provide a new machine-readable layer to the Web where the content of Web pages is annotated with structured data (e.g., RDFa [1]). However, the Web in its current form is made up of at least 15 billion Web pages.¹ Most of these websites are unstructured in nature. Realizing the vision of a usable and up-to-date Web of Data thus requires scalable and accurate natural-language-processing approaches that allow extracting RDF from such unstructured data. Three tasks play a central role when extracting RDF from unstructured data: named entity recognition (NER), named entity disambiguation (NED), also known as entity linking [16], and relation extraction (RE). For the first sentence of Example 1, an accurate named entity recognition approach would return the strings Barack Obama and Washington,

¹ Data gathered from <http://www.worldwidewebsize.com/> on January 4th, 2014.

P. Mika et al. (Eds.) ISWC 2014, Part I, LNCS 8796, pp. 457–471, 2014.
© Springer International Publishing Switzerland 2014

21st century



Rethink How Scientific Knowledge is Communicated



*“The lightbulb was **not** invented by improving the candle.”*

Oren Harari

Digitalization is **more** than just Digitization!

Current and future scientific challenges can not be tackled with an outdated communication system.

**Digitalize Knowledge,
Not Documents!**

Example: Requirements Engineering and Empirical Research

Google Scholar

Artikel

Ungefähr 3.970.000 Ergebnisse (0,16 Sek.)

Beliebige Zeit

Seit 2024

Seit 2023

Seit 2020

Zeitraum wählen...

Nach Relevanz sortieren

Nach Datum sortieren

Beliebige Sprache

Seiten auf Deutsch

Alle Typen

Übersichtsarbeiten

☐ Patente einschließen

☒ Zitate einschließen

[HTML] Empirical research in requirements engineering: trends and opportunities

T Ambreen, N Ikram, M Usman, M Niazi - Requirements Engineering, 2018 - Springer

... trends and future **research** directions. To represent a state-of-the-art of **requirements engineering**, along with various trends and opportunities of **empirical RE research**, we conducted a ...

☆ Speichern 99 Zitieren Zitiert von: 125 Ähnliche Artikel Alle 7 Versionen

[HTML] springer.com

[HTML] Empirical research methodologies and studies in Requirements Engineering: How far did we come?

M Daneva, D Damian, A Marchetto, O Pastor - Journal of systems and ..., 2014 - Elsevier

... Software **Engineering** paradigm. We summarize prior **empirical research** in RE and introduce the contributors to this special issue on **empirical research** methodologies and studies in ...

☆ Speichern 99 Zitieren Zitiert von: 67 Ähnliche Artikel Alle 12 Versionen

[HTML] sciencedirect.com

A systematic literature review of empirical research on quality requirements

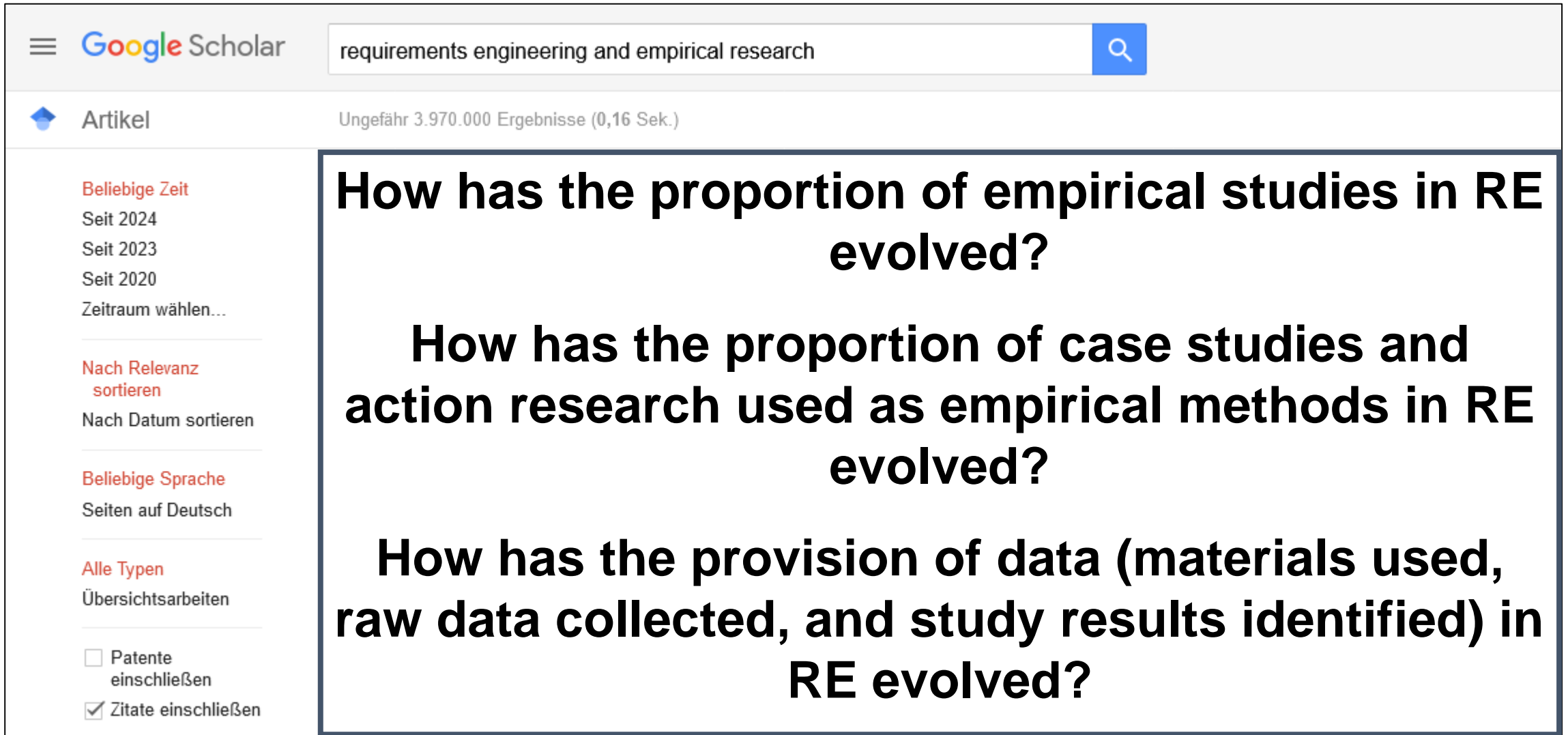
T Olsson, S Sentilles, E Papatheocharous - Requirements Engineering, 2022 - Springer

... We want to understand the **empirical research** on quality **requirements** topics as well as ... We conclude that more **research** is needed as **empirical research** on quality **requirements** is not ...

☆ Speichern 99 Zitieren Zitiert von: 13 Ähnliche Artikel Alle 9 Versionen

[PDF] springer.com

Example: Requirements Engineering and Empirical Research



The image shows a screenshot of a Google Scholar search results page. The search query is "requirements engineering and empirical research". The results are filtered by "Artikel" (Articles) and show approximately 3,970,000 results in 0.16 seconds. The left sidebar contains filters for time (Beliebige Zeit, Seit 2024, Seit 2023, Seit 2020, Zeitraum wählen...), relevance (Nach Relevanz sortieren, Nach Datum sortieren), language (Beliebige Sprache, Seiten auf Deutsch), and type (Alle Typen, Übersichtsarbeiten). There are also checkboxes for "Patente einschließen" and "Zitate einschließen". The main content area displays three research questions in large, bold, black text, each on a new line.

Google Scholar requirements engineering and empirical research

Artikel Ungefähr 3.970.000 Ergebnisse (0,16 Sek.)

How has the proportion of empirical studies in RE evolved?

How has the proportion of case studies and action research used as empirical methods in RE evolved?

How has the provision of data (materials used, raw data collected, and study results identified) in RE evolved?

Beliebige Zeit
Seit 2024
Seit 2023
Seit 2020
Zeitraum wählen...

Nach Relevanz sortieren
Nach Datum sortieren

Beliebige Sprache
Seiten auf Deutsch

Alle Typen
Übersichtsarbeiten

☐ Patente einschließen
☒ Zitate einschließen

How do We Answer These Questions so far?



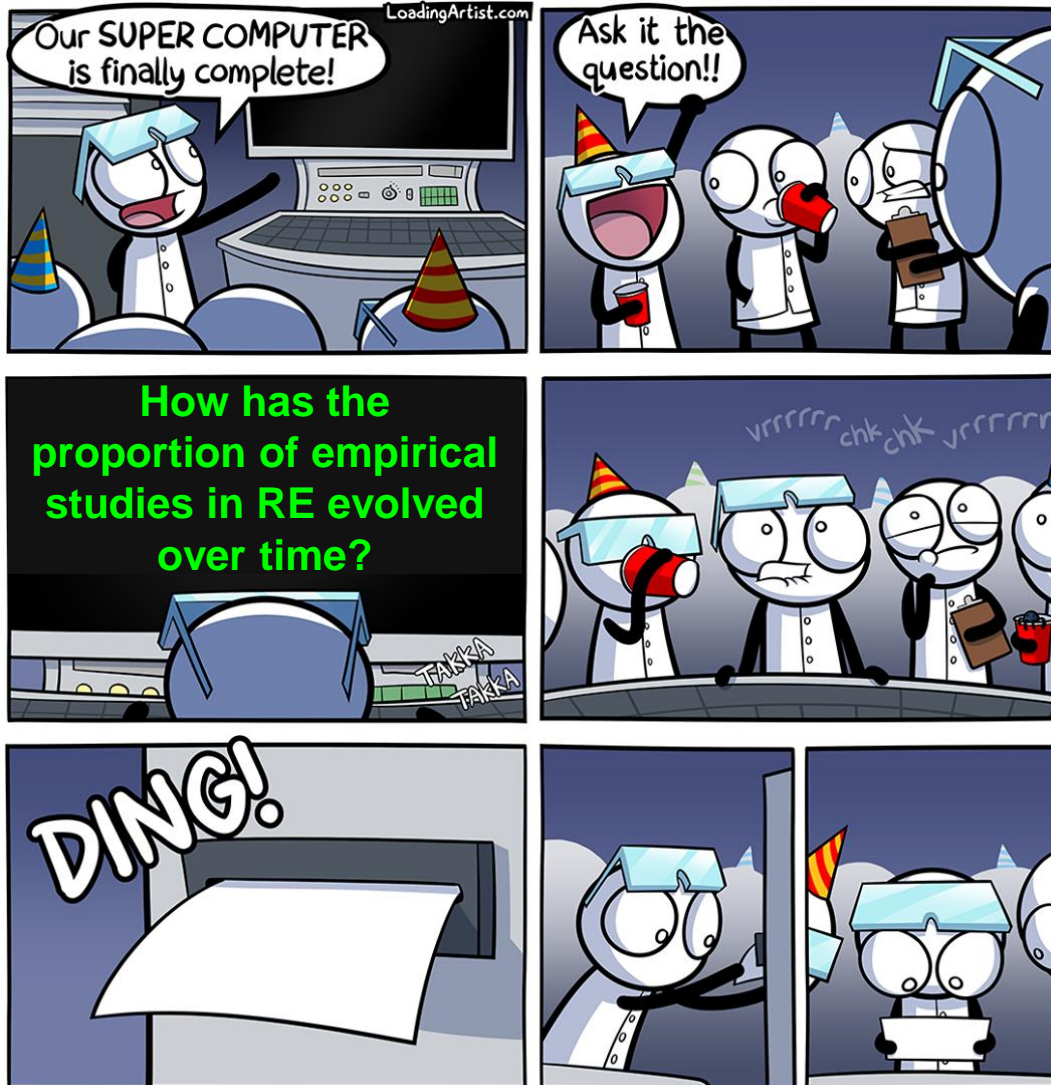
(S)LRs and SMSs for a **comprehensive, up-to-date,**
and **long-term available** overview.



Over **7 million publications** per year^[4]
with an **increasing** (exponential) tendency.

[4] Fire and Guestrin: *Over-Optimization of Academic Publishing Metrics: Observing Goodhart's Law in Action*. GigaScience Vol. 8, No. 6., DOI: [10.1093/gigascience/giz053](https://doi.org/10.1093/gigascience/giz053), 2019.

Wouldn't it be Great if we Could Ask the Computer?



How can we achieve this goal?

Open Research Knowledge Graph (ORKG)

The screenshot displays the ORKG website interface. At the top, there is a navigation bar with the ORKG logo, links for View, Tools, and About, a dropdown menu for NFDI4DataScience, a search bar, and buttons for '+ Add new' and 'Sign in'. Below the navigation bar, a large banner features the text 'Scholarly Knowledge. FAIR.' and a description of the ORKG's purpose: 'The Open Research Knowledge Graph (ORKG) aims to describe research papers in a structured manner. With the ORKG, papers are easier to find and compare.' A 'Play video' button is also present. The main content area is titled 'Browse by research field' and includes a search bar for fields. Below this, five red buttons represent different research fields: Arts and Humanities (432 papers - 33 comparisons), Engineering (3391 papers - 345 comparisons), Life Sciences (4142 papers - 195 comparisons), Physical Sciences & Mathematics (15694 papers - 731 comparisons), and Social and Behavioral Sciences (844 papers - 170 comparisons). The 'Comparisons' tab is selected, showing a list of research papers. The first entry is 'Systematic Literature Review (SLR) Tools analysed based on General Features' in the 'Information Science' field, dated 04-03-2024. The second entry is 'RO estimates for infectious diseases' in the 'Virology' field, dated 04-04-2024. The third entry is 'Machine learning of pre-harvesting crop/fruit parameters to minimize overall losses in farming production' in the 'Plant Cultivation, PL...' field. On the right side, there are three sections: 'Load Tools' with a note about the Mastodon widget, 'ORKG stories' with a link to 'Find out more', and 'Join ORKG!' with a 'Sign up' button. At the bottom right, there is a section titled 'Knowledge base for science' with a note about the increasing volume of research publications.

ORKG

View Tools About NFDI4DataScience

Search... + Add new Sign in

Scholarly Knowledge. FAIR.

The Open Research Knowledge Graph (ORKG) aims to describe research papers in a structured manner. With the ORKG, papers are easier to find and compare. [Play video](#)

Browse by research field

Search for fields...

Arts and Humanities
432 papers - 33 comparisons

Engineering
3391 papers - 345 comparisons

Life Sciences
4142 papers - 195 comparisons

Physical Sciences & Mathematics
15694 papers - 731 comparisons

Social and Behavioral Sciences
844 papers - 170 comparisons

Comparisons Papers Visualizations Reviews Lists

Top recent

Systematic Literature Review (SLR) Tools analysed based on General Features
16 Contributions 0 Visualizations 04-03-2024
The non-AI characteristics of SLR tools analyzed based on 23 features in the paper titled "Artificial Intelligence for Literature Reviews: Opportunities and Challenges"

RO estimates for infectious diseases
6 Contributions 0 Visualizations 04-04-2024
RO estimates for infectious diseases

Machine learning of pre-harvesting crop/fruit parameters to minimize overall losses in farming production

Load Tools By loading the Mastodon widget, you agree with the [cookie guidelines](#)

ORKG stories
See how researchers benefit from using ORKG.
[Find out more](#)

Join ORKG! [Sign up](#)

Knowledge base for science
We are flooded with new publications in research every day and it is increasingly challenging to keep up

The ORKG **revolutionizes** how **scientific knowledge** is communicated and processed, making it **actionable** for **machines** and **researchers** to navigate, compare, and review **vast amounts of information** efficiently.

Research Knowledge Graphs

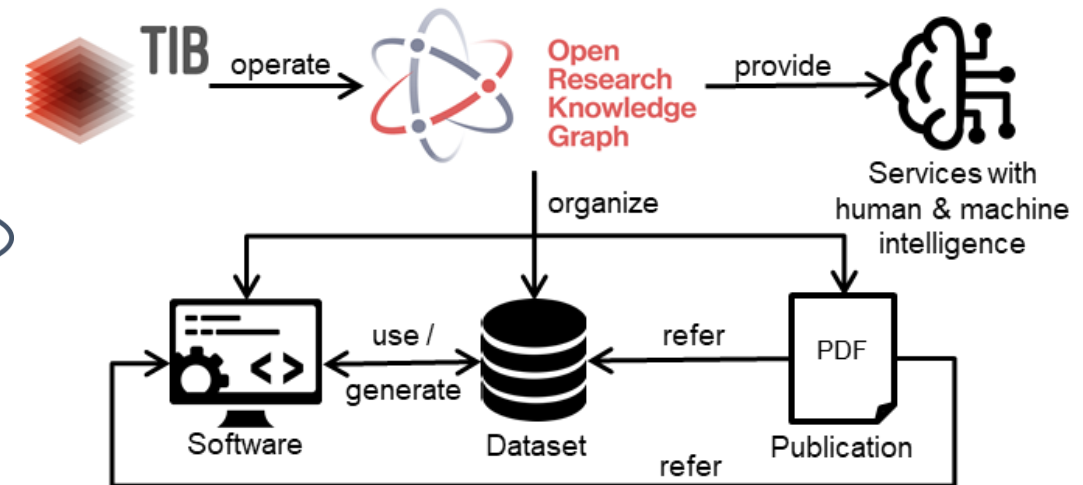
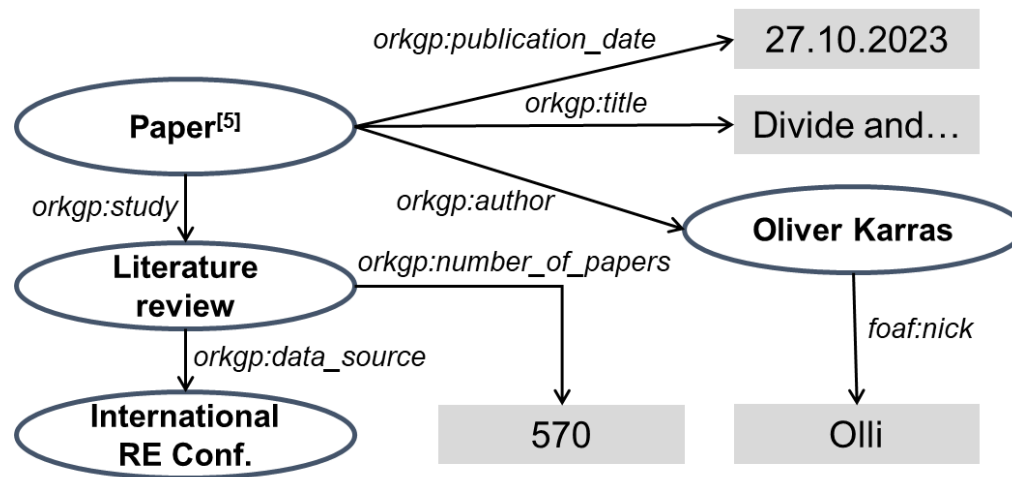
Research Knowledge Graphs (RKGs) are a **technology** for organizing scientific (meta-)data in a

- **Flexible, fine-grained, and semantic** representation
- That is **understandable** and **processable** by humans and machines

Research Knowledge Graphs

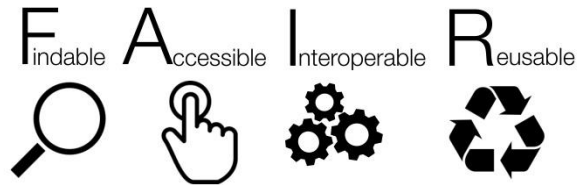
Research Knowledge Graphs (RKGs) are a **technology** for organizing scientific (meta-)data in a

- **Flexible, fine-grained, and semantic** representation
- That is **understandable** and **processable** by humans and machines



The **ORKG** is a **ready-to-use** and **sustainably** operated **infrastructure** with **services** that uses a **cross-discipline RKG** for the **long-term** and **openly available organization** of scientific (meta-)data according to the **FAIR** data principles.

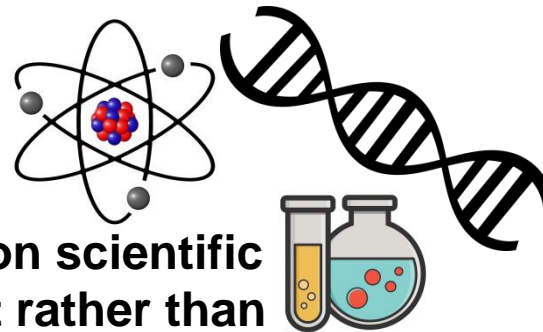
ORKG Objectives



Make research FAIR



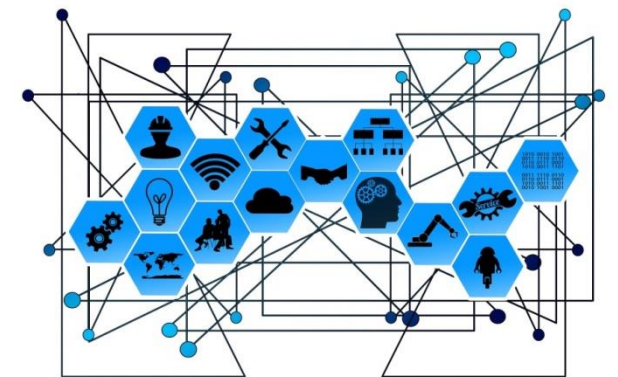
**Provide long-term
available and up-to-date
overviews**



**Focus on scientific
content rather than
pure metadata**



Foster collaboration



**Tackle interdisciplinary
challenges**

 Assign to observatory

Behind the Scenes

Mining User Requirements from Application Store Reviews Using Frame Semantics

Nishant Jha and Anas Mahmoud^(✉)

The Division of Computer Science and Engineering, Louisiana State University,
Baton Rouge, LA 70803, USA
njha1@lsu.edu, mahmoud@csc.lsu.edu

Abstract. *Context and motivation:* Research on mining user reviews in mobile application (app) stores has noticeably advanced in the past few years. The majority of the proposed techniques rely on classifying the textual description of user reviews into different categories of technically informative user requirements and uninformative feedback. *Question/Problem:* Relying on the textual attributes of reviews often produces high dimensional models. This increases the complexity of the classifier and can lead to overfitting problems. *Principal ideas/results:* We propose a novel semantic approach for app review classification. The proposed approach is based on the notion of semantic role labeling, or characterizing the lexical meaning of text in terms of semantic frames. Semantic frames help to generalize from text (individual words) to more abstract scenarios (contexts). This reduces the dimensionality of the data and enhances the predictive capabilities of the classifier. Three datasets of user reviews are used to conduct our experimental analysis. Results show that semantic frames can be used to generate lower dimensional and more accurate models in comparison to text classification methods. *Contribution:* A novel semantic approach for extracting user requirements from app reviews. The proposed approach enables a more efficient classification process and reduces the chance of overfitting.

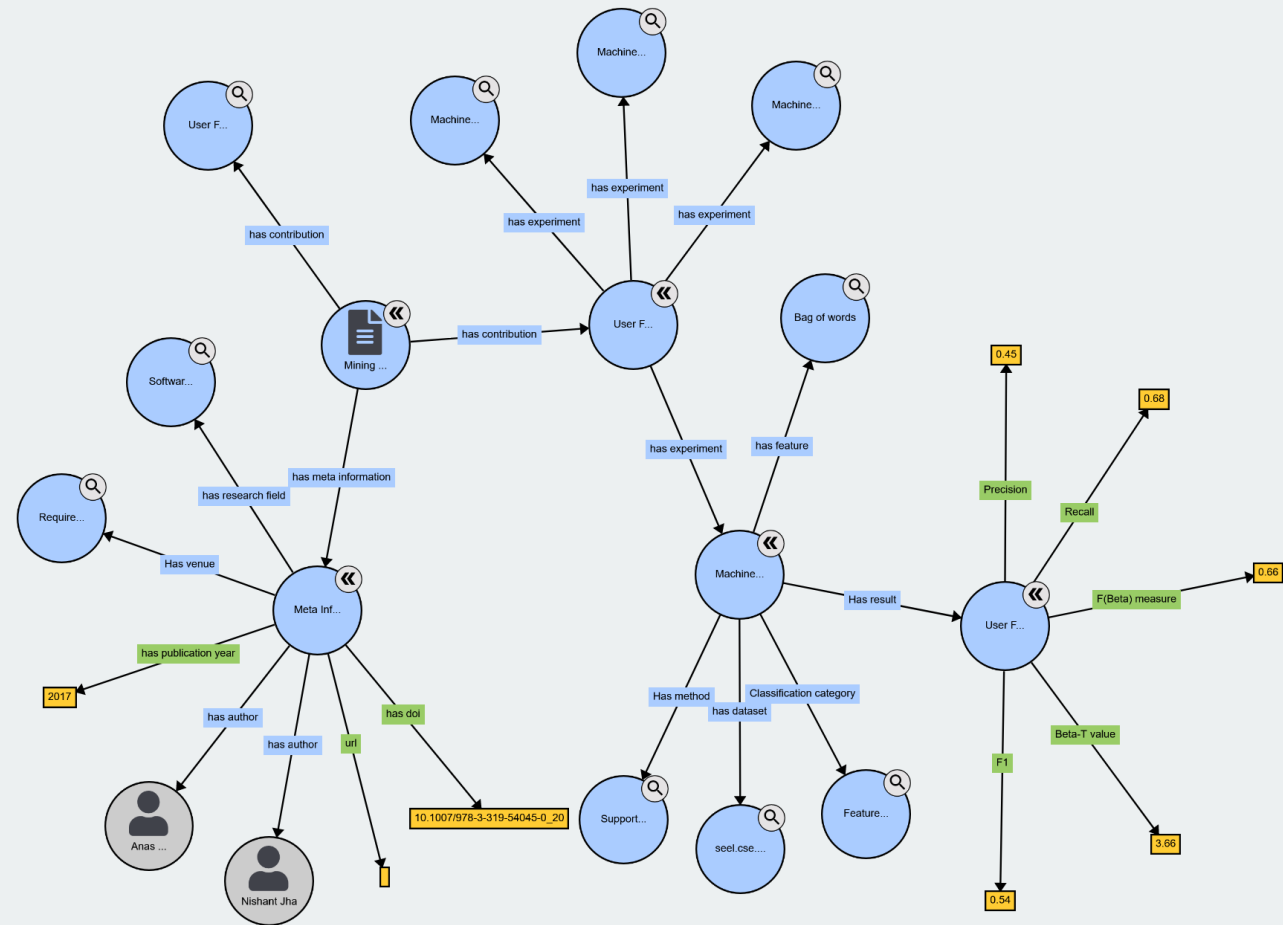
Keywords: Requirements elicitation · Application stores · Classification

1 Introduction

Mobile application markets, or app stores (e.g., Google Play and Apple App Store), represent a unique model of service-oriented business. Such platforms have created an unprecedented opportunity for app developers to directly monitor the opinions of a large population of end-users of their software [25]. Through app stores feedback services, app users can directly share their experience in the form of textual reviews and meta-data (e.g., star ratings). Analyzing large datasets of app store reviews has revealed that they contain substantial amounts of up-to-date technical information. Such information can be leveraged by app developers to help them maintain and sustain their apps in a highly-competitive

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P. Grünbacher and A. Perini (Eds.): REFSQ 2017, LNCS 10153, pp. 273–287, 2017.
DOI: 10.1007/978-3-319-54045-0_20

Mining User Requirements from Application Store Reviews Using Frame Semantics



Using FAIR Scientific Knowledge

How has the provision of data (materials used, raw data collected, and study results identified) in RE evolved?

Mining User Requirements from Application Store Reviews Using Frame Semantics

Nishant Jha and Anas Mahmoud^(✉)

The Division of Computer Science and Engineering, Louisiana State University,
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njha1@lsu.edu, mahmoud@csc.lsu.edu

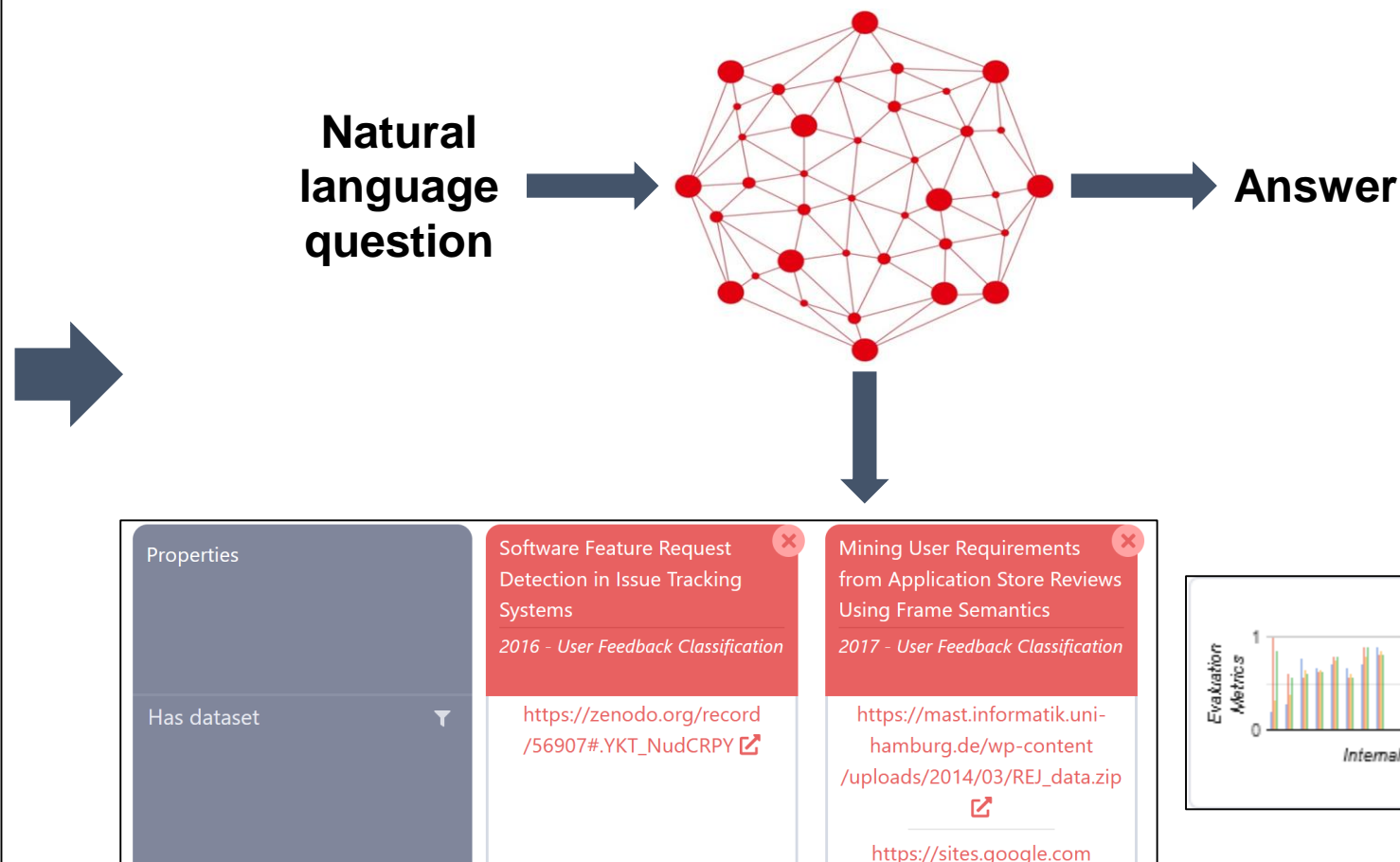
Abstract. *Context and motivation:* Research on mining user reviews in mobile application (app) stores has noticeably advanced in the past few years. The majority of the proposed techniques rely on classifying the textual description of user reviews into different categories of technically informative user requirements and uninformative feedback. *Question/Problem:* Relying on the textual attributes of reviews often produces high dimensional models. This increases the complexity of the classifier and can lead to overfitting problems. *Principal ideas/results:* We propose a novel semantic approach for app review classification. The proposed approach is based on the notion of semantic role labeling, or characterizing the lexical meaning of text in terms of semantic frames. Semantic frames help to generalize from text (individual words) to more abstract scenarios (contexts). This reduces the dimensionality of the data and enhances the predictive capabilities of the classifier. Three datasets of user reviews are used to conduct our experimental analysis. Results show that semantic frames can be used to generate lower dimensional and more accurate models in comparison to text classification methods. *Contribution:* A novel semantic approach for extracting user requirements from app reviews. The proposed approach enables a more efficient classification process and reduces the chance of overfitting.

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P. Grünbacher and A. Perini (Eds.): REFSQ 2017, LNCS 10153, pp. 273–287, 2017.
DOI: 10.1007/978-3-319-54045-0_20



ORKG Comparisons

Acknowledgement
of creators

DOI

Visualizations

Interactive filtering

Overview of Approaches that Classify User Feedback as Feature Request ☆👁

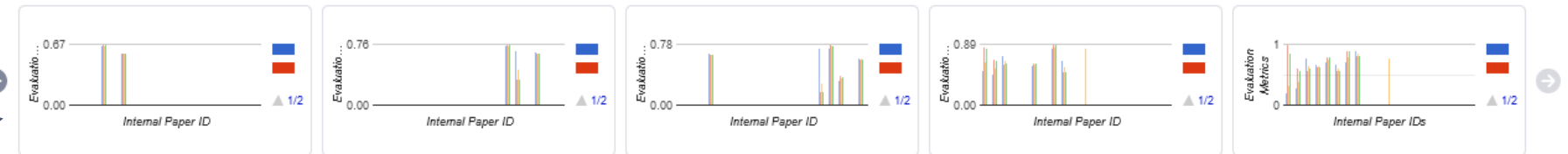
📅 June 2021

👤 Oliver Karras

👤 Eduard C. Groen

This overview shows the classification results of approaches that use the machine learning algorithms Naïve Bayes, Support Vector Machines, and Decision Trees C4.5 in combination with the machine learning features Bag of Words or Term Frequency - Inverse Document Frequency to classify user feedback as feature request.

DOI: <https://doi.org/10.48366/r112387>



Properties

Software Feature Request
Detection in Issue Tracking
Systems

User Feedback Classification - 2016

Mining User Requirements from
Application Store Reviews Using
Frame Semantics

User Feedback Classification - 2017

Mining Twitter Feeds for Software
User Requirements

User Feedback Classification - 2017

Automatic Classification of Non-
Functional Requirements from
Augmented App User Reviews

User Feedback Classification - 2017

Bug reports
simply p

classifying

User Feedback

has dataset

https://zenodo.org/record/56907#.YKT_NudCRPY

https://mast.informatik.uni-hamburg.de/wp-content/uploads/2014/03/REJ_data.zip

<https://sites.google.com/site/appuserreviews/>
seel.cse.lsu.edu/data/refsq17.zip

seel.cse.lsu.edu/data/re17.zip

Not available

https://mast.informatik.uni-hamburg.de/wp-content/uploads/2014/03/REJ_data.zip

ORKG Comparisons are Citable

Comparison of Studies on Germany's Energy Supply in 2050

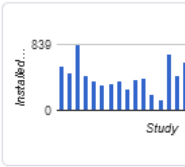
November 2021

Felix Kullmann Jan Göpfert Oliver Karras Patrick Kuckertz Sören Auer Markus Stocker

Peter Markewitz Leander Kotzur Detlef Stolten

This comparison compares the results of various studies analyzing a future low-carbon energy system for Germany. The focus of this study comparison is electricity generation. In the future, however, other essential characteristics of the respective energy system designs in the individual studies will be listed. Installed capacity is given in GW and electricity generation is given in TWh.

DOI: <https://doi.org/10.21203/rs.3.rs-1000000/v1>



Comparison of studies on Germany's energy supply in 2050

Autoren Felix Kullmann, Peter Markewitz, Detlef Stolten, Oliver Karras, Patrick Kuckertz, Leander Kotzur, Jan-Maris Göpfert, Sören Auer, Markus Stocker


Publikationsdatum 2021

Ausgabe FZJ-2022-00782

Verlag Technoökonomische Systemanalyse





Beschreibung This comparison compiles the results from various studies analyzing a future low-carbon energy system for Germany. The focus of this study comparison is electricity generation. In the future, however, other essential characteristics of the respective energy system designs in the individual studies will be listed. Installed capacity is given in GW and electricity generation is given in TWh.

Zitate insgesamt **Zitiert von: 2**

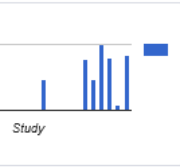


Google Scholar-Artikel [Comparison of studies on Germany's energy supply in 2050](#)
F Kullmann, P Markewitz, D Stolten, O Karras... - 2021
[Zitiert von: 2](#) [Ähnliche Artikel](#)

Share

NFDI4ing
Energy System
Research



Properties

[has energy sources](#)

Klimaneutralität

Contributions

Den Weg zu einem treibhausgasneutralen Deutschland ressourcenschonend gestalten

Contribution 1 - 2019

[all sources](#)

[bioenergy](#)

[geothermics](#)

[hydropower](#)

[import](#)

[net import](#)

net import

net import

net import

net import

net import

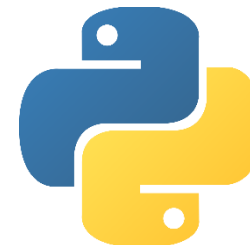
So far so Good, but...

...what can we do with machine-actionable scientific knowledge?

Anything we want!

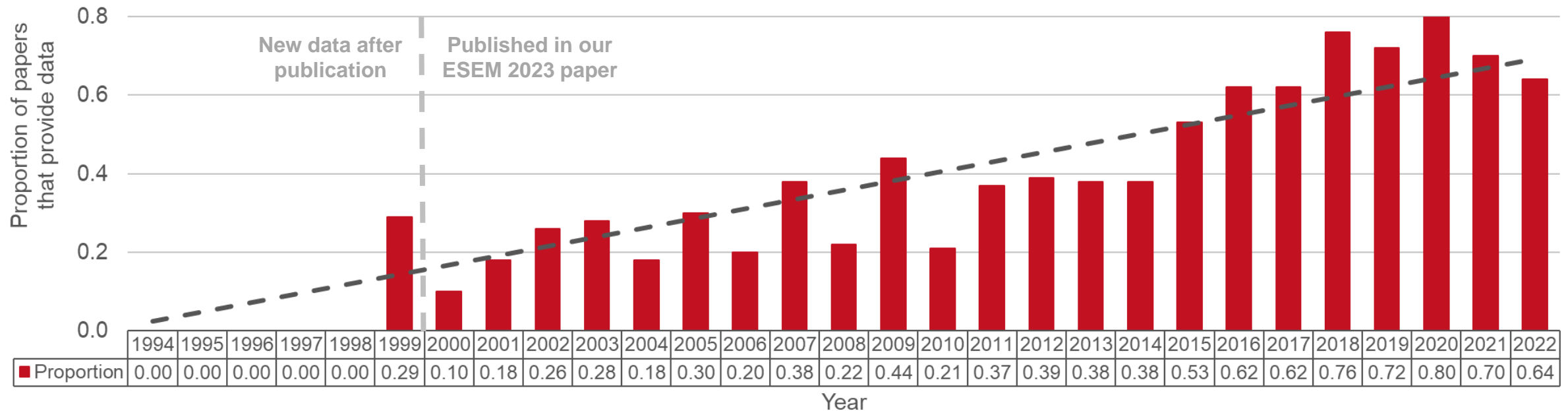
1. Papers, comparisons, and visualizations in the ORKG are **openly available for (re-)use and extension to anyone.**
2. The ORKG provides several **interfaces** for processing the data, e.g., to develop novel search, retrieval, mining, and assistance applications.

{ REST : API }



(Re-)use and Extension of Scientific Knowledge

How has the provision of data (materials used, raw data collected, and study results identified) in RE evolved?^[5, 6, 7]
(Based on 680 papers in the ORKG)



[5] Karras et al.: *Divide and Conquer the EmpiRE: A Community-Maintainable Knowledge Graph of Empirical Research in Requirements Engineering*. 2023 ACM/IEEE International Symposium on Empirical Software Engineering and Measurement (ESEM), DOI: [10.1109/ESEM56168.2023.10304795](https://doi.org/10.1109/ESEM56168.2023.10304795), 2023.

[6] Project on GitHub: <https://github.com/okarras/EmpiRE-Analysis>

[7] Interactive Jupyter notebook: <https://mybinder.org/v2/gh/okarras/EmpiRE-Analysis/HEAD?labpath=%2Fempire-analysis.ipynb>

ORKG Observatory

Empirical Software Engineering | Observatory

Edit

This observatory works to provide a community-maintainable knowledge graph of empirical research in software engineering. Our goal is to continuously acquire and curate comprehensive knowledge about empirical research applied in scientific publications in the research field of software engineering and its subfields, such as requirements engineering. In this way, we want to provide a comprehensive, up-to-date, and long-term overview of the state-of-the-art on empirical research in software engineering.

We are currently working on a knowledge graph of empirical research in requirements engineering.


For this purpose, we have developed a corresponding ORKG template (cf. <https://orkg.org/template/R186491>).

Contact: oliver.karras@tib.eu


Research problems

- empirical research in software engineering
- empirical research in requirements engineering

Organizations




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Universität
Hannover




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FOR SCIENCE AND TECHNOLOGY
UNIVERSITY LIBRARY

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Oliver Karras
TIB - Leibniz Information Centre for Science and
Technology



Jil Kluender
Leibniz University Hannover

Content | 688 items

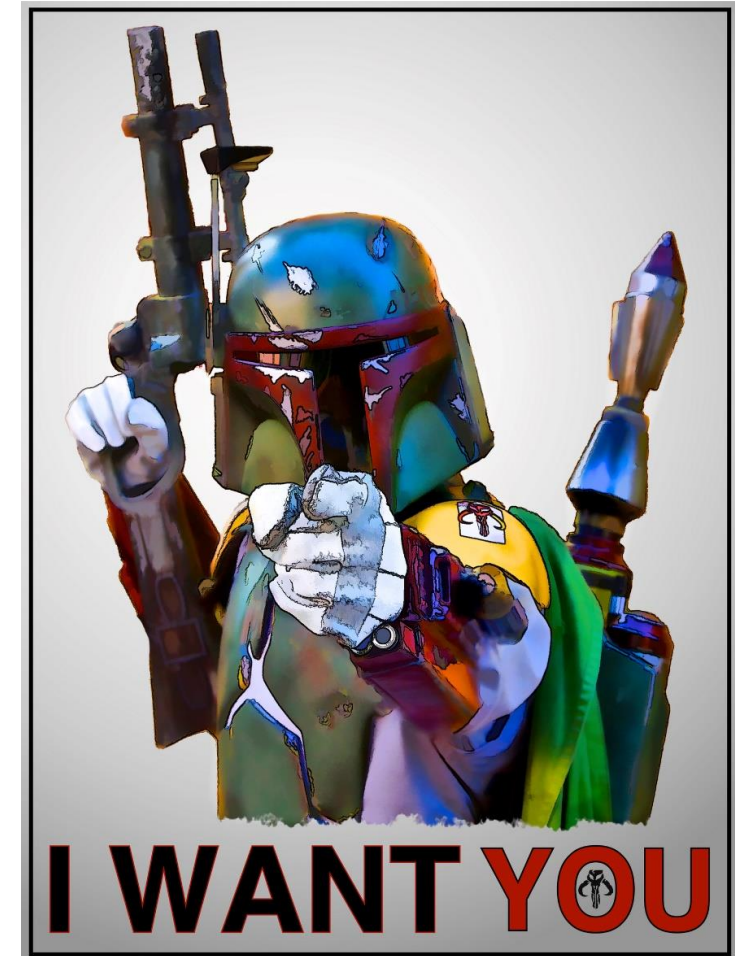
Show: ☒ Paper ☒ Comparison ☒ Visualization Top recent

- **Open** groups maintain topics in the ORKG
- **Central** access point for the community to all curated contents

Who organizes this scientific knowledge in the ORKG?

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The heart of the ORKG is its **CROWD**:
Researchers from any discipline!



How can We Contribute to the ORKG?

1. While writing a paper:

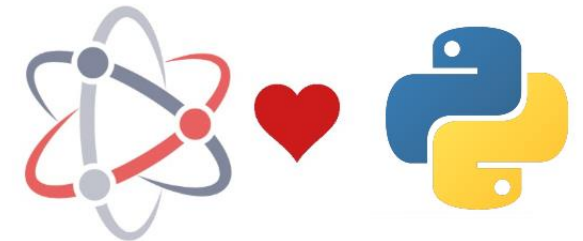
- SciKGT_{EX}: LaTeX package for FAIR annotations in a paper (embedded in the PDF) that can be imported into ORKG
- [https://github.com/Christof93/SciKGT_{EX}](https://github.com/Christof93/SciKGTex)



SciKGT_{EX}

2. While developing an analysis:

- Python package (& R package in development) for FAIR annotations in analysis scripts that can be imported into the ORKG
- <https://pypi.org/project/orkg/>



3. At any time:

- Manually using the ORKG Frontend to describe papers
- (Semi-)automatically using the ORKG REST API



Open
Research
Knowledge
Graph

{ REST : API }

What We Will Learn to Contribute to the ORKG

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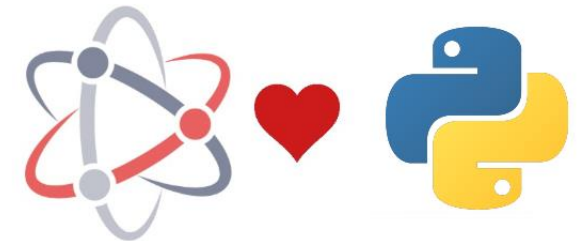
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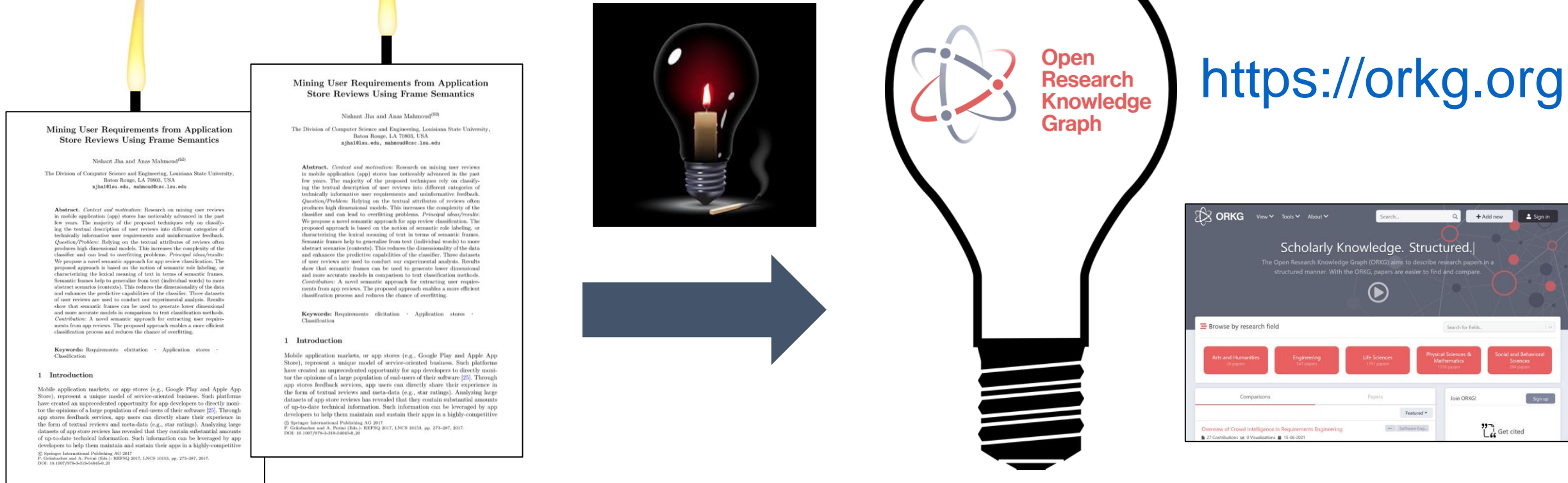
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Conclusion



Let's bring scholarly communication and open science in Requirements Engineering to the 21st century!

Further Reading



Open Access ORKG Book

- Celebrating the 5th anniversary of ORKG
- A practical guide for new and advanced users
 - ORKG's terms and concepts
 - ORKG's approach
 - ORKG's technology
 - ORKG's success stories

<https://cuvillier.de/de/shop/publications/9037-open-research-knowledge-graph>

